

---

# Integrated Modeling of Cognition and the Information Environment



**Michael D. Byrne**  
Department of Psychology  
Rice University



**Alex Kirlik**  
Aviation Human Factors  
University of Illinois

# Overview

- ◆ Approach
- ◆ Model Structure
  - Cognitive model
  - Physical and environmental models
- ◆ Error behavior
  - Coverage and sources
  - Answers to questions
- ◆ Work in progress/future extensions



# General Approach

- ◆ Traditional cognitive modeling approaches
  - History of modeling simple, static laboratory tasks
  - Now ready to handle complex, dynamic environments
  - How?
- ◆ Traditional ecological approaches
  - Good for describing task-environmental structure
  - Make simplistic assumptions about the operator
- ◆ Our goal: Unify the two approaches
  - Cognitive model informed by environmental analysis

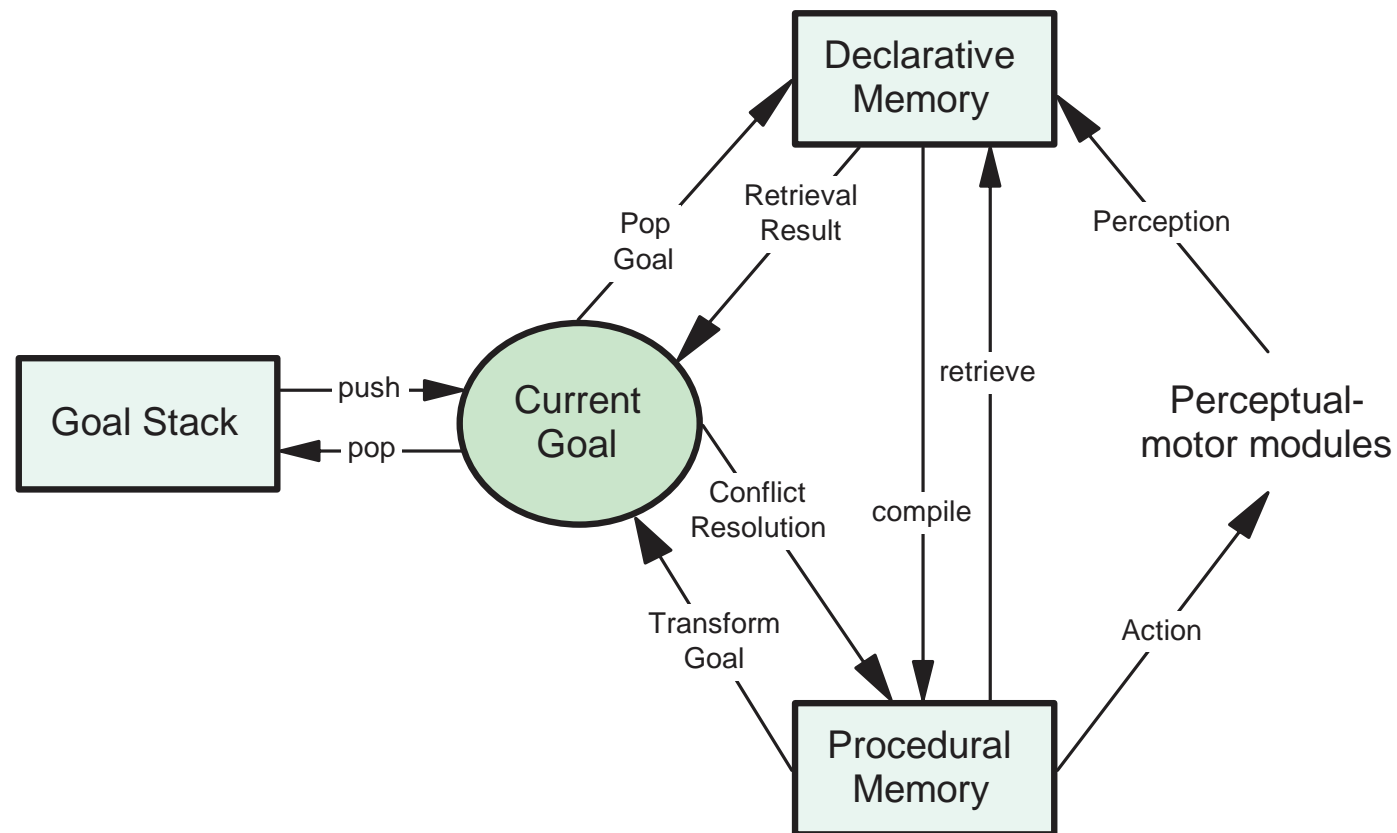


# ACT-R

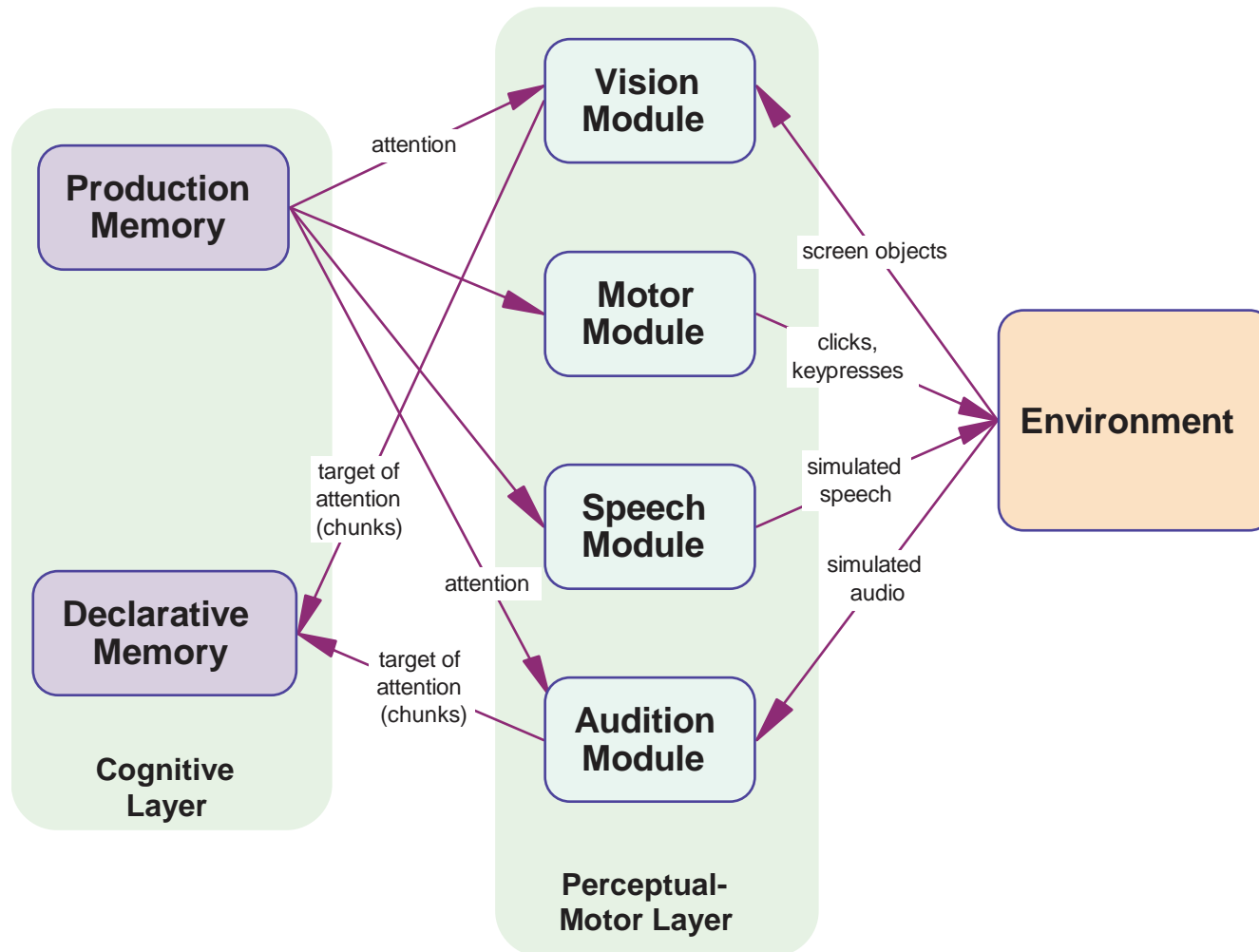
- ◆ ACT-R computational cognitive architecture
  - Production system
  - Semantic network
- ◆ Based on “rational analysis”
  - Activation of items in the semantic network driven by a Bayesian equation combining current system context with frequency & recency information
    - Activation determines retrieval probability and speed
  - Production selection (called “conflict resolution”) driven by equation balancing goal value, cost (in time), and success rate
- ◆ Important note: System is noisy



# ACT-R



# ACT-R/PM



# Ecological/Task Analysis

- ◆ Use environmental modeling to provide the ACT-R/PM model a realistic “external” environment
  - For example, realistic time constraints based on model of aircraft dynamics, runway layout, information layout, etc.
- ◆ Use environmental analysis (based in part on SMEs) to:
  - Identify problem-solving and decision-making strategies
  - Set parameters in ACT-R representing the information landscape for those strategies
    - Frequency and recency
    - Success rate and costs



# Model Scope

- ◆ Model of single individual, the pilot, and the environment
- ◆ Currently, we do not model the FO
- ◆ Also, no model of errors resulting from miscommunications between agents
  - Not presently a major strength of ACT-R, and it appeared to us that other models could better address this
- ◆ Does not model low-level control of steering
  - Airport is a series of “rails”
  - However, steering g-force constraints respected

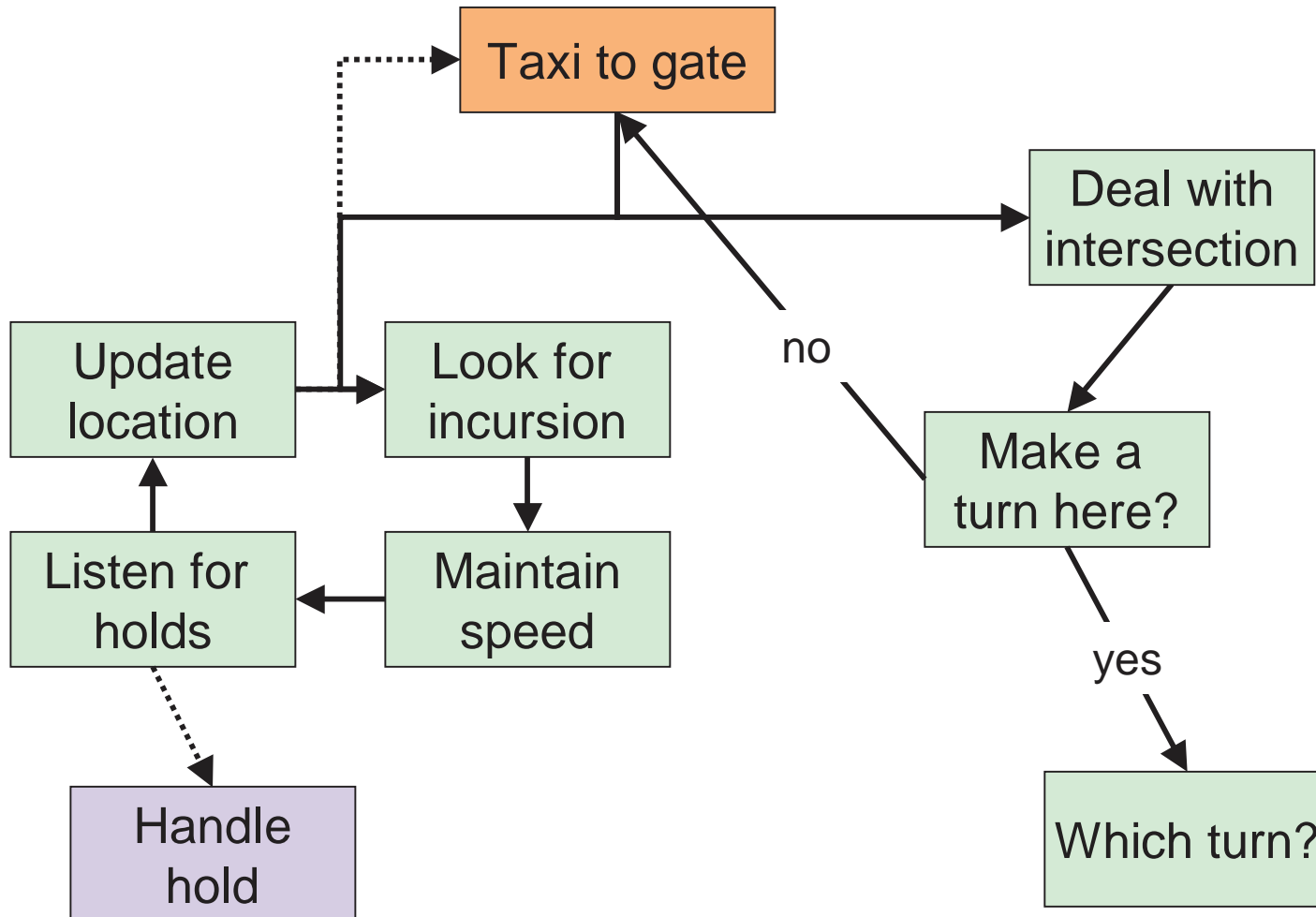
*Focus on Adaptation and Cognitive Limits to Adaptation*

---





# Decision (Goal Selection) Flow



# Maintenance Goals

- ◆ During routine (straight) taxiing, all these goals will regularly be made the focus
- ◆ When one of these goals completes, it can return information to the top goal
  - Example 1: If an incursion is detected, it will return a note to the main goal to next push a goal to handle the incursion
  - Example 2: Updating location might determine that there's an intersection coming up, which will return a note to the main goal to deal with it
- ◆ Satisfying these goals takes time



# Look for Incursion

- ◆ Visual scan of scene looking for anything untoward on a rail
- ◆ Will pick up other objects that may be relevant, like new signs in view
  - If no incursion, then this will be returned to the top goal
- ◆ If there is an incursion, top goal is told so
- ◆ Top goal pushes “handle incursion” subgoal
  - Behavior would be to break as quickly as possible/necessary
  - Not actually implemented



# Listen for Hold

- ◆ Very rapid in the case of no available auditory stimuli
- ◆ When such stimuli are available, listen for a few moments to determine if this is a hold issuance
- ◆ If so, return to top goal with that information
- ◆ Top goal pushes “deal with hold” subgoal



# Maintain Speed

- ◆ If the model did low-level steering, this would be more inclusive
- ◆ Checks speed against standard speed bounds
- ◆ If plane is too fast, either back off throttle or apply brake
- ◆ If plane is too slow, either let up on brake or increase throttle
- ◆ Fairly rapid, but there is a little time in there to actually make the decision and to perform the relevant motor movements



# Update Location

- ◆ Current location represented in a qualitative way
  - On taxiway X
  - Between taxiways Y1 and Y2
  - Heading toward Y2
- ◆ Updated primarily by reference to signs
- ◆ In a richer visual environment, this would be much more developed
  - Visual scene cues (especially in familiar airports)
  - Radio cues



# Make a Turn Here?

- ◆ This can be very simple:
  - If the intersection coming up is a “T” then a turn must be made
  - Otherwise, model generally relies on memory of turns to decide whether to turn
    - Expectancies can play a role here
- ◆ This is a potential error source
  - Makeup of errors suggests that this is uncommon as a decision error, though can easily happen as a planning error



# Which Turn?

- ◆ Model explicitly chooses a strategy for determining which turn to make
- ◆ Different strategies have different time demands
- ◆ Thus, model is sensitive to environmental constraints
  - Aircraft dynamics
  - Sign placement
  - Taxiway geometry
- ◆ Considers time cost and rough success rate information
  - Most accurate strategy given time available (e.g. Payne, et al.)





# Turn Decision Strategies

- ◆ Strategies available:
  - Remember
    - Fast, increasingly inaccurate
  - Turn toward gate
    - Not quite as fast, surprisingly accurate in most airports
  - Turn which reduces larger of XY distance
    - Moderately fast, much more accurate than you'd think
  - Derive from “map knowledge”
    - Slow
    - High accuracy in principle, but still error-prone
- ◆ Buy time and re-assess (brake)

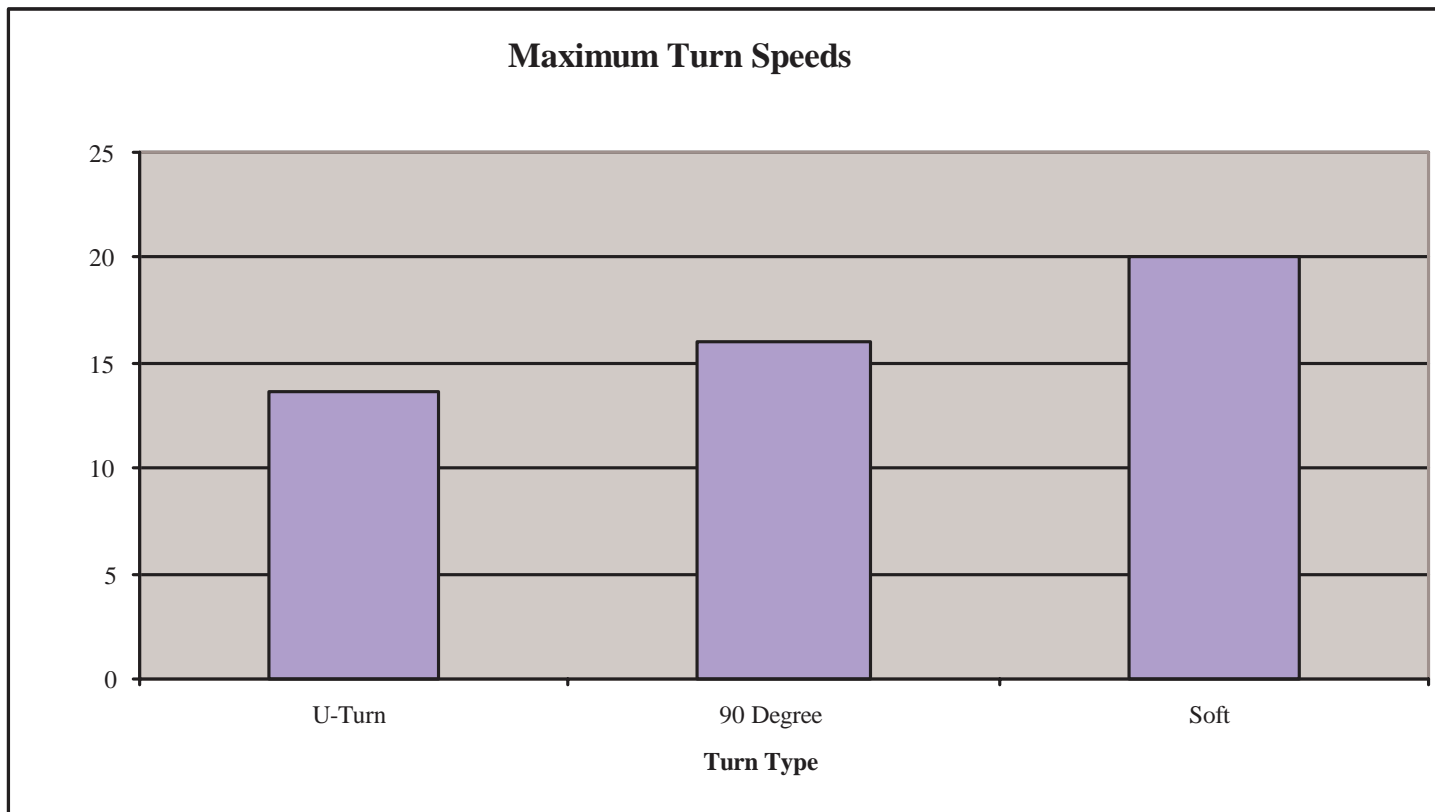


# Turn Execution

- ◆ Speed in a turn is determined by
  - Turn radius (hard, 90, soft)
  - G-force limitations (guideline is 0.25 g's)
  - Model brakes a/c in time to meet speed threshold
- ◆ While we don't model the control movements made by the pilot during the turn, we assume that this requires visual guidance
  - We “lock” the visual system to the relevant yellow line during the turn



# Max g Turn Analysis by Type

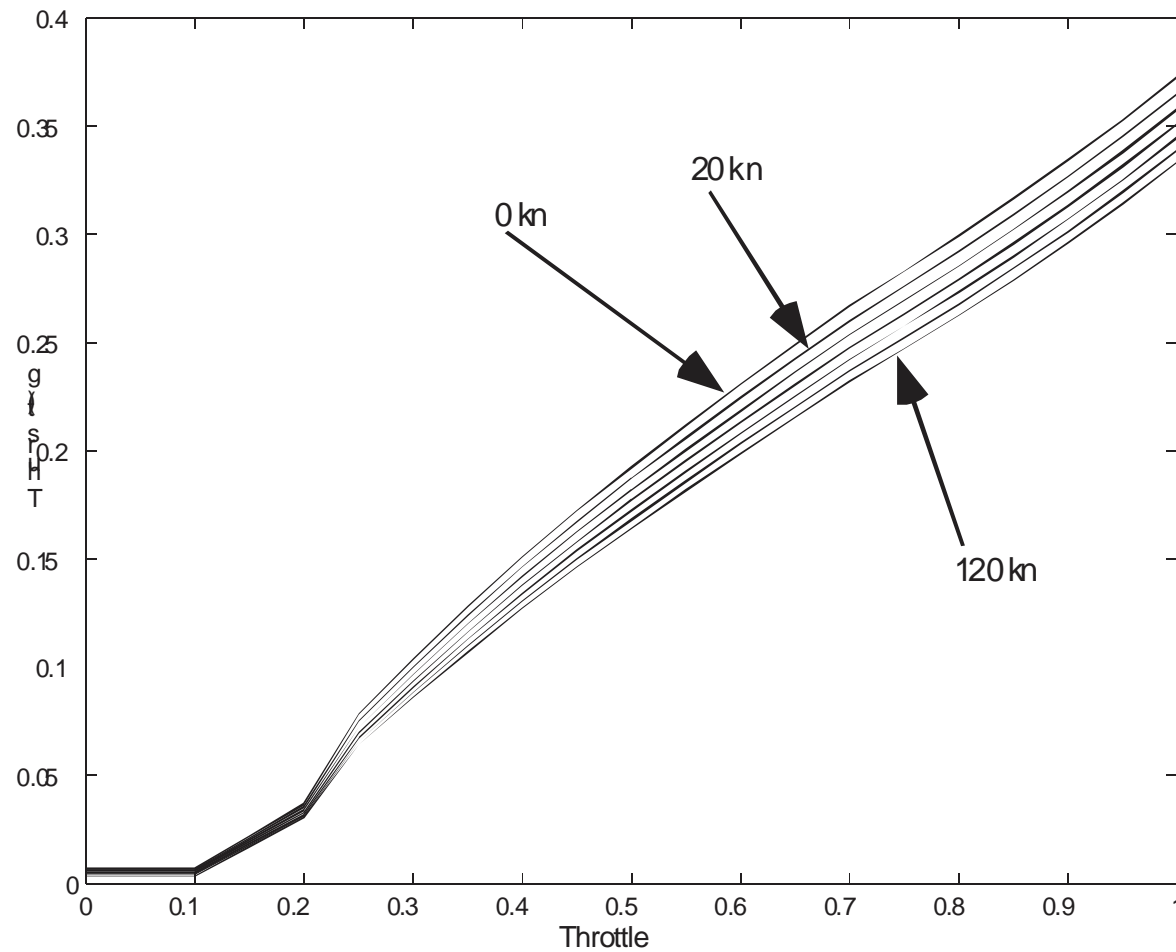


# Physical Model

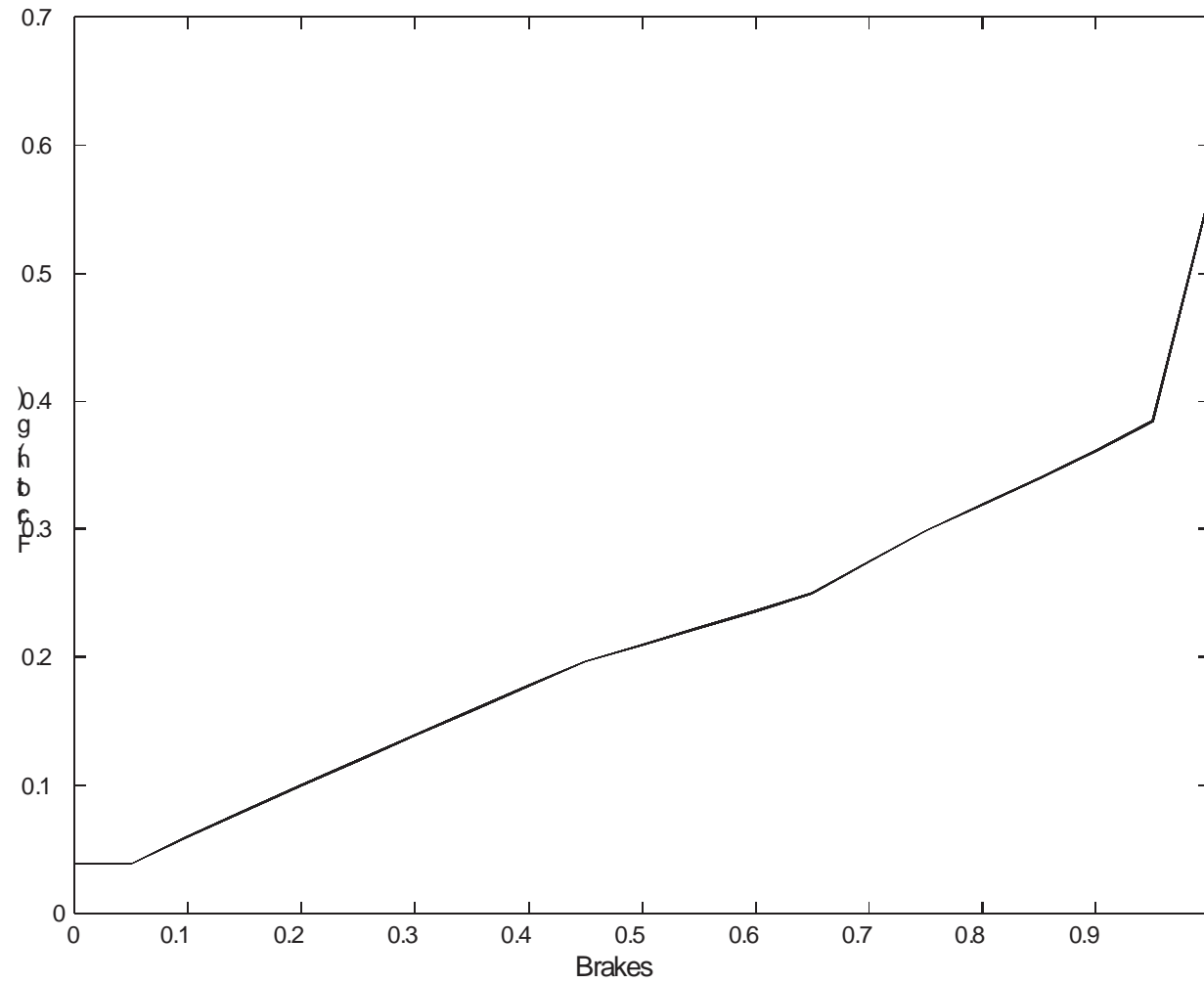
- ◆ Model of physical aircraft based on
  - Nissan car simulator
  - Aircraft specifications from Boeing and NASA
  - Adjustments from physics first principles
- ◆ This model determines aircraft response to
  - Thrust
  - Braking
- ◆ Time is a crucial resource to the cognitive model --  
Physical model provides temporal “landscape”



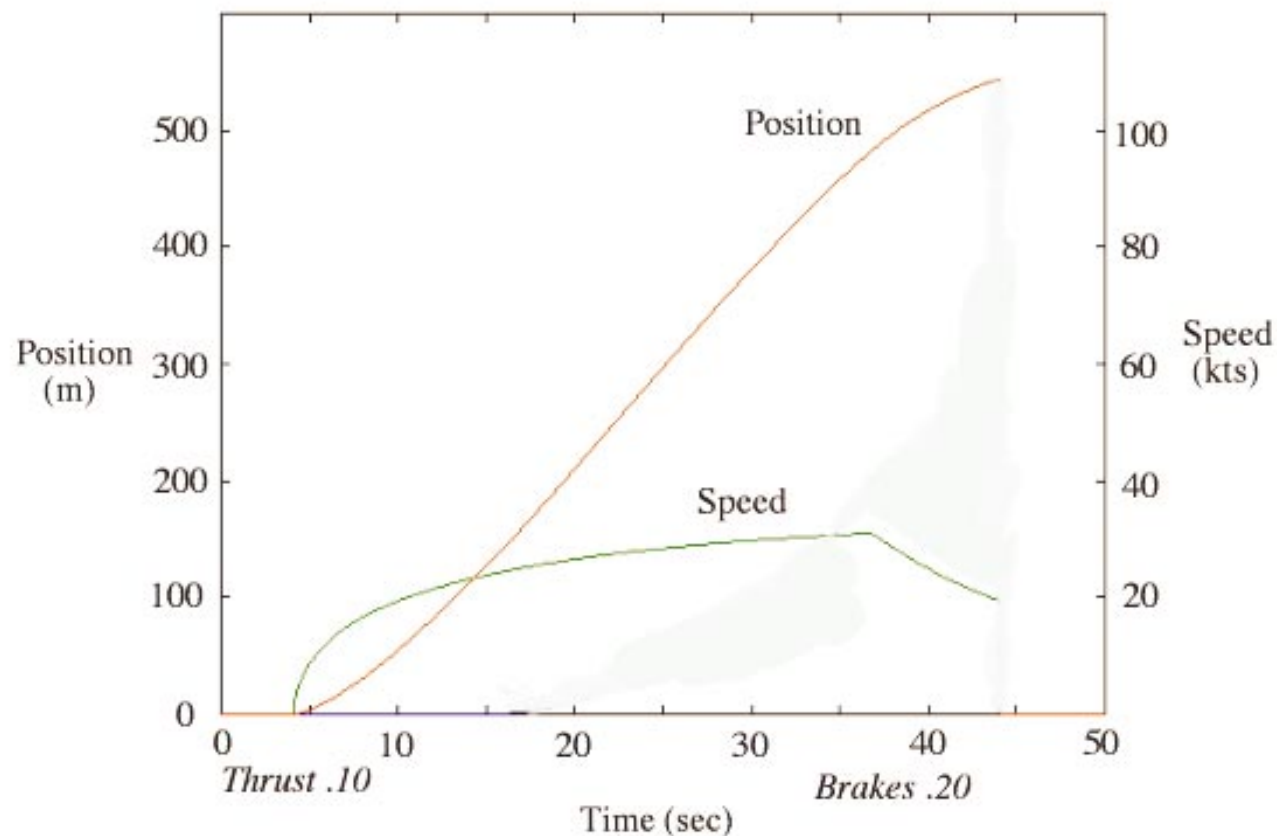
# Thrust Effects -Cheng et al. 2001



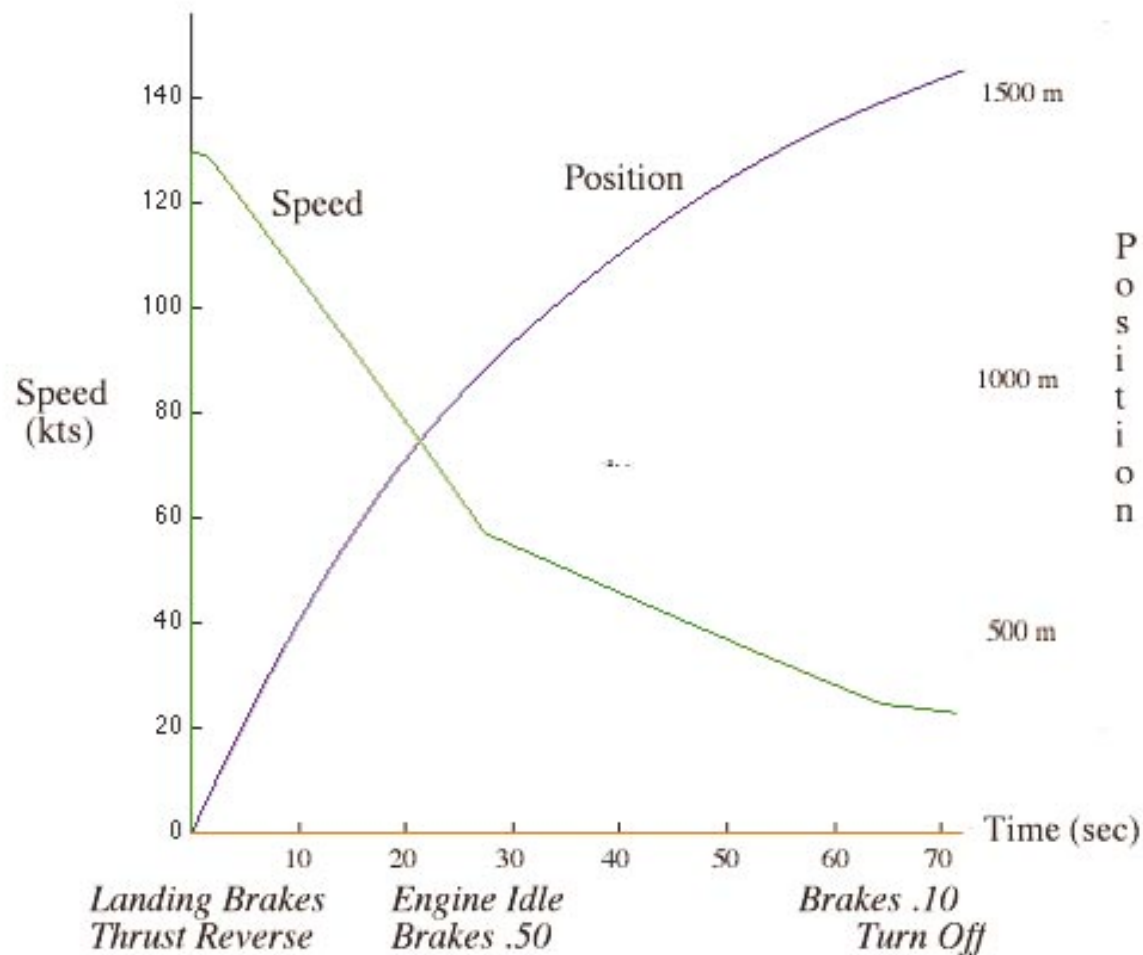
# Brake Effects - Cheng et al. 2001



# Aircraft Model: Start from Stop



# A/C Model: Landing Timeline





# Visual Environment Model

- ◆ Used the database from the NASA flight simulator
- ◆ Aircraft position and heading used to determine what objects should be visible
  - Yellow lines
  - Signs
  - Distance from each
- ◆ Work is in progress on degrading the representation of text at longer distances
  - ACT-R/PM's Vision Module contains a “best guess” mechanism for degraded input
  - This is another potential error source

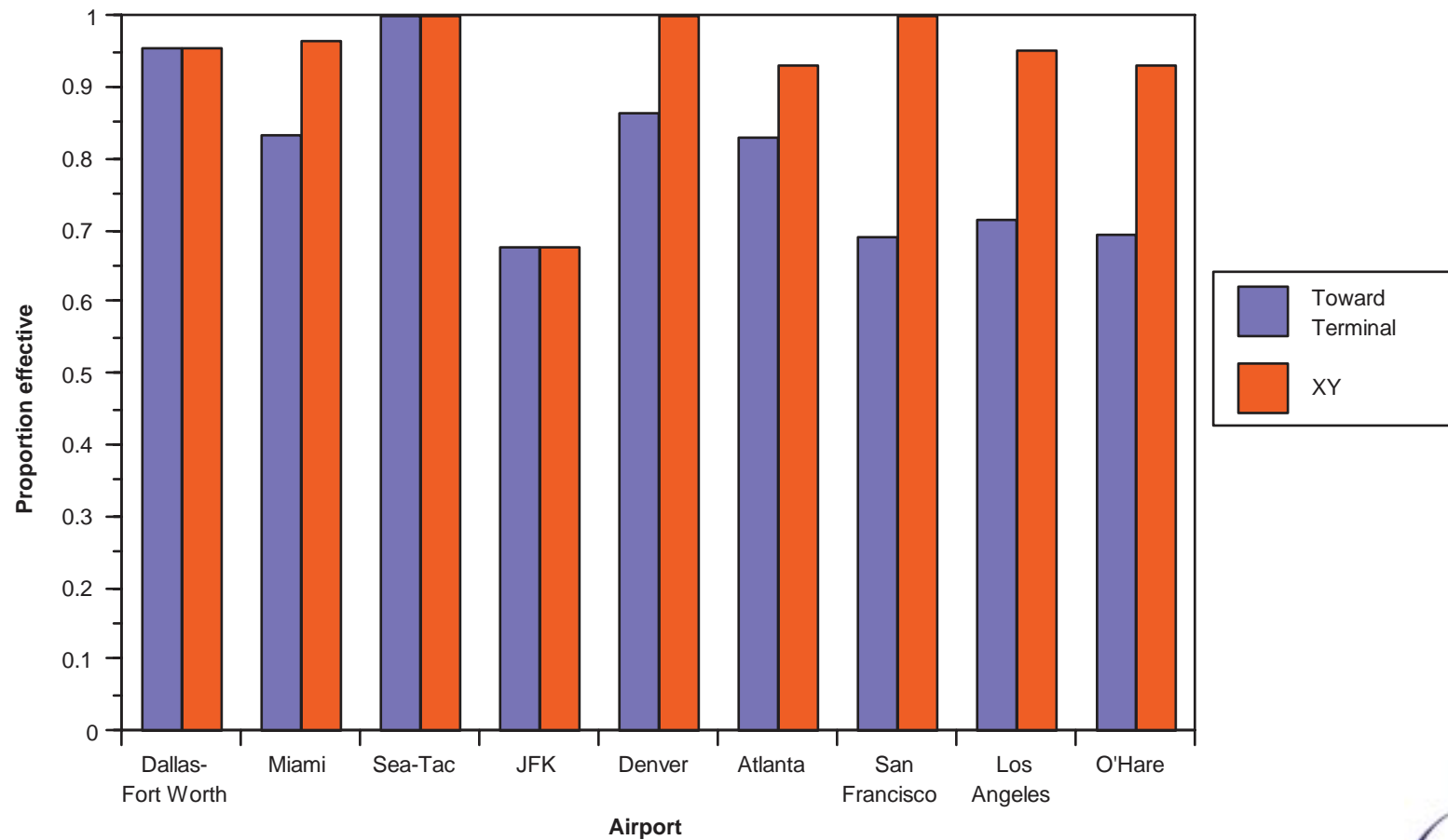


# Task Environment Model

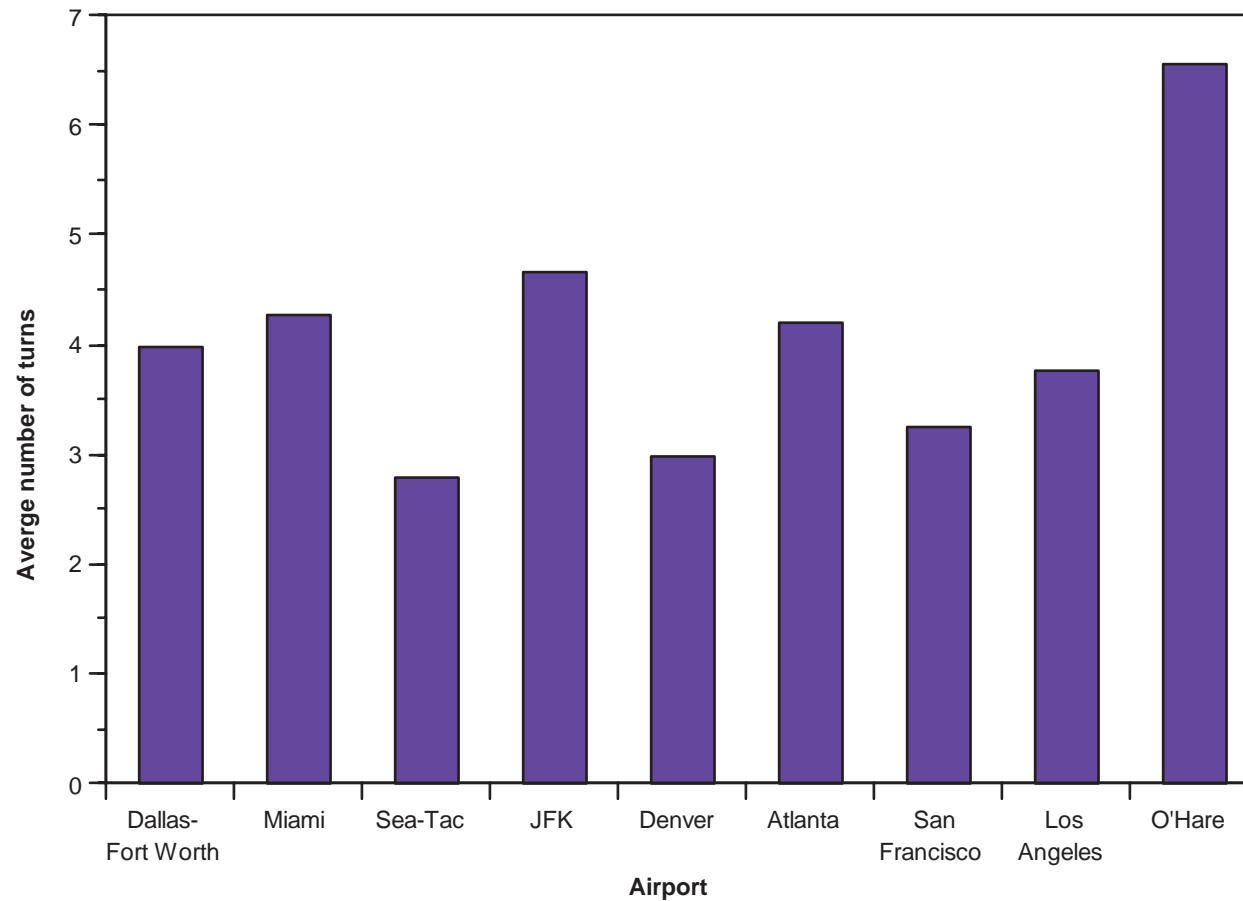
- ◆ SME provided us with Jepp charts for other airports with “nominal” taxi routes indicated. N = 284 total turns analyzed
  - Different airports
    - Near grids: Atlanta, Dallas/Ft Worth, SeaTac, Denver
    - More like O’Hare: JFK
    - In between: San Francisco, Miami, Los Angeles
- ◆ Discoveries:
  - “XY” heuristic is good across the board
  - “Toward terminal” heuristic is good many places, but not at O’Hare
  - All simulated turns at O’Hare where *both* these heuristics fail, at least one error was made ....



# Heuristic Effectiveness



# Average Taxi Route Length



# Error Behavior

- ◆ Several sources
  - Retrieval failure/mis-retrieval
    - Exacerbated by memory-based workload
  - Use of less accurate strategies to meet time constraints
    - Exacerbated by temporal workload
  - Perceptual failures
- ◆ Coverage
  - The decision errors are at least amenable to explanation
  - Prediction is difficult
    - Need a priori basis for setting all parameters for all pilots
  - Some execution errors can be modeled



# Continuing/Future Work

- ◆ Very near term
  - Monte Carlo simulations to explore parameter sensitivity
- ◆ Already mentioned
  - Degraded perceptual inputs
- ◆ Questions to answer
  - Are there other decision strategies? If so, how long do they take and how well do they work?
- ◆ Adding FO model
  - Would need more detailed information about FO tasks to help determine behavior of that model



# Questions

- ◆ Is there a way to validate the conclusions from the modeling?
  - Would need more data, and more detailed data
  - Might be able to test some of the model's tendencies better by more closely examining model's behavior and designing studies that really test where model is most vulnerable to error
  - More time for model-building wouldn't hurt



# Other credits

- ◆ Brian Webster
- ◆ Michael Fleetwood
- ◆ Chris Fick
- ◆ NASA

